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Pinning control design for feedback stabilization of constrained Boolean control networks

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Abstract

We study the pinning control design for feedback stabilization of constrained Boolean control networks (BCNs) via the semitensor product of matrices. Firstly, a constrained algebraic representation is obtained for constrained BCNs with pinning control, and a necessary and sufficient condition is established for the reachability of constrained BCNs with pinning control. Secondly, a general procedure is proposed for the pinning control design of state feedback stabilization of constrained BCNs. Thirdly, a necessary and sufficient condition is presented for the output feedback stabilization of constrained BCNs with pinning control. Finally, two illustrative examples are worked out to demonstrate the effectiveness of the obtained new results.

MSC: 93C55; 93B52

Keywords: Boolean control network; pinning control; feedback stabilization; constrained control; semitensor product of matrices

1 Introduction

In recent years, with the development of systems biology and medical science, there have been a lot of computational models to simulate gene regulatory networks (GRNs) [1, 2]. In 1969, Kauffman [3] firstly proposed the concept of Boolean networks to model GRNs. From then on, the study of Boolean networks has attracted attention of many scholars [4–7]. In a Boolean network, genes are simulated as 1 and 0 for studying the activity of genes. Since the dynamics of Boolean networks does not contain parameters, we can use Boolean networks to model large-scale GRNs.

Recently, a semitensor product method [8] has been proposed for the study of Boolean networks. Using this novel method, we can convert a Boolean (control) network into a linear (bilinear) form. Then, we can conveniently investigate Boolean networks with the help of classical control theory. This framework is called the algebraic state space representation (ASSR) of logical systems. Up to now, many scholars have applied the ASSR framework to the analysis and control of Boolean networks and obtained lots of interesting results [9–23]. Moreover, the semitensor product method has also been applied to networked evolutionary games [24, 25] and feedback shift registers [26].

As one of the most important issues in the study of GRNs, the stabilization of Boolean control networks (BCNs) has been found to be widely applied in the design of therapeutic

interventions and the explanation of some living phenomena [27]. The state feedback stabilization of BCNs was studied in [28, 29], and a novel design procedure was established. Later, the design of output feedback stabilizers of BCNs was investigated [30–32], and some necessary and sufficient conditions were presented.

Note that the pinning control problem has been introduced to the study of Boolean networks in some recent works [33, 34]. With the pinning control, we can realize the control objective by controlling a small number of nodes. Using the ASSR framework, the pinning control design for the stabilization of Boolean networks was studied in [34], and a novel design procedure was established. Moreover, in GRNs, some states and inputs are undesirable because they may lead to a dangerous situation such as the deterioration of a disease, the metastasis of a cancer [7, 35], and so on. Hence, we need put some constraints to the undesirable states and inputs in BCNs. It should be noticed that the pinning control design for feedback stabilization of constrained BCNs is still a challenging problem, and there are fewer results on this problem. In addition, the pinning stabilization control design procedure proposed in [34] cannot be directly applied to the pinning control design for feedback stabilization of constrained BCNs. This motivates us to study the pinning control design for feedback stabilization of constrained BCNs.

In this paper, using the ASSR framework, we investigate the pinning control design for feedback stabilization of constrained BCNs and present a number of new results. The main contributions of this paper are as follows. On one hand, we convert the dynamics of constrained BCNs with pinning control into a constrained algebraic form, which facilitates the reachability analysis and control design of constrained BCNs with pinning control. On the other hand, we present some necessary and sufficient conditions for the pinning control design of feedback stabilizers of constrained BCNs, which can be easily verified with the help of MATLAB.

The rest of this paper is organized as follows. In Section 2, we give some preliminaries on the semitensor product of matrices. In Section 3, we investigate the pinning control design for feedback stabilization of constrained BCNs and present the main results of this paper. Two illustrative examples are worked out to verify the obtained new results in Section 4, which is followed by a brief conclusion in Section 5.

Notation \mathbb{R} , \mathbb{N} , and \mathbb{Z}_+ denote the sets of real numbers, natural numbers, and positive integers, respectively; $\mathcal{D} := \{1, 0\}$, $\mathcal{D}^n := \underbrace{\mathcal{D} \times \dots \times \mathcal{D}}_n$, $\Delta_n := \{\delta_n^k \mid 1 \leq k \leq n\}$, where δ_n^k denotes the k th column of I_n , and for compactness, $\Delta := \Delta_2$. An $n \times t$ matrix M is called a logical matrix if $M = [\delta_n^{i_1} \delta_n^{i_2} \dots \delta_n^{i_t}]$. We express M briefly as $M = \delta_n[i_1 \ i_2 \ \dots \ i_t]$. Denote the set of $n \times t$ logical matrices by $\mathcal{L}_{n \times t}$. An $n \times t$ matrix $M = (M_{ij})$ is called a Boolean matrix if $M_{ij} \in \mathcal{D}$, $i = 1, \dots, n$, $j = 1, \dots, t$. Denote the set of $n \times t$ Boolean matrices by $\mathcal{B}_{n \times t}$. Given a real matrix $A \in \mathbb{R}^{m \times n}$, $\text{Col}_i(A)$ denotes the i th column of A , and $\text{Row}_i(A)$ denotes the i th row of A ; ‘ \ast ’ denotes the Khatri-Rao product of matrices.

2 Preliminaries

First of all, we give some fundamental preliminaries on the semitensor product of matrices. For details, we refer to [8].

Definition 1 ([8]) The semitensor product of two matrices $A \in \mathbb{R}^{m \times n}$ and $B \in \mathbb{R}^{p \times q}$ is

$$A \times B = (A \otimes I_n^{\frac{q}{p}})(B \otimes I_p^{\frac{n}{q}}), \tag{1}$$

where $\alpha = \text{lcm}(n, p)$ represents the least common multiple of n and p , and \otimes is the Kronecker product.

Obviously, when $n = p$, the semitensor product of A and B becomes the conventional matrix product. Thus, we can simply call it ‘product’ and omit the symbol ‘ \ltimes ’ if no confusion arises.

Proposition 1 ([8]) *Let $X \in \mathbb{R}^{t \times 1}$ be a column vector, and $A \in \mathbb{R}^{m \times n}$. Then*

$$X \ltimes A = (I_t \otimes A) \ltimes X. \tag{2}$$

Proposition 2 ([8]) *Let $X \in \mathbb{R}^{m \times 1}$ and $Y \in \mathbb{R}^{n \times 1}$ be two column vectors. Then*

$$Y \ltimes X = W_{[m,n]} \ltimes X \ltimes Y, \tag{3}$$

where $W_{[m,n]} \in \mathcal{L}_{mn \times mn}$ is the so-called swap matrix defined as

$$W_{[m,n]} = \delta_{mn} \begin{bmatrix} 1 & m+1 & \cdots & (n-1)m+1 \\ & 2 & m+2 & \cdots & (n-1)m+2 \\ & & & \cdots & \\ & & & & m & m+m & \cdots & (n-1)m+m \end{bmatrix}.$$

Proposition 3 ([8]) *Define the matrix, $M_{r,k}$, called the k -valued power-reducing matrix, as*

$$M_{r,k} = \begin{bmatrix} \delta_k^1 & 0_k & \cdots & 0_k \\ 0_k & \delta_k^2 & \cdots & 0_k \\ \vdots & & & \\ 0_k & 0_k & \cdots & \delta_k^k \end{bmatrix}, \tag{4}$$

where $0_k \in \mathbb{R}^k$ is the zero vector.

Denoting $1 \sim \delta_2^1$ and $0 \sim \delta_2^2$, we have $\Delta \sim \mathcal{D}$, where ‘ \sim ’ denotes two different forms of the same object. In most places of this paper, we use δ_2^1 and δ_2^2 to express logical variables and call them the vector form of logical variables. The following lemma is fundamental for the matrix expression of logical functions.

Lemma 1 ([8]) *Let $f(x_1, x_2, \dots, x_s) : \mathcal{D}^s \mapsto \mathcal{D}$ be a logical function. Then, there exists a unique matrix $M_f \in \mathcal{L}_{2 \times 2^s}$, called the structural matrix of f , such that*

$$f(x_1, x_2, \dots, x_s) = M_f \ltimes x_1 \ltimes \cdots \ltimes x_s := M_f \ltimes_{i=1}^s x_i, \quad x_i \in \Delta. \tag{5}$$

For example, the structural matrices for negation (\neg), conjunction (\wedge), disjunction (\vee), conditional (\rightarrow), biconditional (\leftrightarrow), and exclusive Or ($\bar{\vee}$) are $M_n = \delta_2[2 \ 1]$, $M_c = \delta_2[1 \ 2 \ 2 \ 2]$, $M_d = \delta_2[1 \ 1 \ 1 \ 2]$, $M_i = \delta_2[1 \ 2 \ 1 \ 1]$, $M_e = \delta_2[1 \ 2 \ 2 \ 1]$ and $M_p = \delta_2[2 \ 1 \ 1 \ 2]$, respectively.

Based on Lemma 1, we can convert a BCN into an algebraic representation. For details, refer to [8].

3 Main results

In this section, we give the main results of this paper. Firstly, we convert the dynamics of a constrained Boolean network with pinning control into its equivalent constrained algebraic form, based on which we obtain a necessary and sufficient condition for the reachability of constrained BCNs with pinning control. Secondly, we present the pinning control design procedure for the feedback stabilization of constrained BCNs.

3.1 Constrained algebraic form

A Boolean network with n network nodes and r pinning controls can be described as

$$\begin{cases} x_1(t+1) = f_1(x_1(t), \dots, x_n(t), u_1(t)), \\ \vdots \\ x_r(t+1) = f_r(x_1(t), \dots, x_n(t), u_r(t)), \\ x_{r+1}(t+1) = f_{r+1}(x_1(t), \dots, x_n(t)), \\ \vdots \\ x_n(t+1) = f_n(x_1(t), \dots, x_n(t)), \\ y_i(t) = h_i(x_1(t), \dots, x_n(t)), \quad i = 1, \dots, p, \end{cases} \tag{6}$$

where nodes i_1, \dots, i_r are selected to be pinning controlled, and $1 \leq r < n$. Without loss of generality, we assume that $i_s = s, s = 1, \dots, r, x_i(t) \in \mathcal{D}, i = 1, \dots, n, u_i(t) \in \mathcal{D}, i = 1, \dots, r,$ and $y_i(t) \in \mathcal{D}, i = 1, \dots, p,$ are the states, the control inputs, and the outputs of system (6), respectively, and $f_i : \mathcal{D}^{n+1} \mapsto \mathcal{D}, i = 1, \dots, r, f_i : \mathcal{D}^n \mapsto \mathcal{D}, i = r + 1, \dots, n,$ and $h_i : \mathcal{D}^n \mapsto \mathcal{D}, i = 1, \dots, p,$ are logical functions.

In order to convert system (6) into an algebraic form, we define $x(t) = \times_{i=1}^n x_i(t) \in \Delta_{2^n}, x^1(t) = \times_{i=1}^r x_i(t) \in \Delta_{2^r}, x^2(t) = \times_{i=r+1}^n x_i(t) \in \Delta_{2^{n-r}}, u(t) = \times_{i=1}^r u_i(t) \in \Delta_{2^r},$ and $y(t) = \times_{i=1}^p y_i(t) \in \Delta_{2^p}.$ Assume that the structural matrix of f_i and h_i are $F_i, i = 1, \dots, n,$ and $H_i, i = 1, \dots, p,$ respectively. Using Lemma 1, system (6) can be expressed as

$$\begin{aligned} x^1(t+1) &= F_1 x(t) u_1(t) F_2 x(t) u_2(t) \cdots F_r x(t) u_r(t) \\ &= F_1 (I_{2^{n+1}} \otimes F_2) x(t) u_1(t) x(t) u_2(t) F_3 x(t) u_3(t) \cdots F_r x(t) u_r(t) \\ &= F_1 (I_{2^{n+1}} \otimes F_2) W_{[2^n, 2^{n+1}]} M_{r, 2^n} x(t) u_1(t) u_2(t) \\ &\quad \times F_3 x(t) u_3(t) \cdots F_r x(t) u_r(t) \\ &= F_1 \prod_{i=2}^r [(I_{2^{n+i-1}} \otimes F_i) W_{[2^n, 2^{n+i-1}]} M_{r, 2^n}] x(t) u(t) \\ &:= L_1 x(t) u(t), \end{aligned} \tag{7}$$

$$\begin{aligned} x^2(t+1) &= F_{r+1} x(t) F_{r+2} x(t) \cdots F_n x(t) \\ &= F_{r+1} * F_{r+2} * \cdots * F_n x(t) \\ &:= L_2 x(t), \end{aligned} \tag{8}$$

and

$$\begin{aligned}
 y(t) &= H_1x(t)H_2x(t)\cdots H_px(t) \\
 &= H_1 * H_2 * \cdots * H_px(t) \\
 &:= Hx(t),
 \end{aligned}
 \tag{9}$$

where

$$\begin{aligned}
 L_1 &= F_1 \prod_{i=2}^r [(I_{2^{n+i-1}} \otimes F_i)W_{[2^n, 2^{n+i-1}]}M_{r, 2^n}] \in \mathcal{L}_{2^r \times 2^{n+r}}, \\
 M_{r, 2^n} &= \text{diag}\{\delta_{2^n}^1, \delta_{2^n}^2, \dots, \delta_{2^n}^{2^n}\} \in \mathcal{L}_{2^{2n} \times 2^{2n}}, \\
 L_2 &= F_{r+1} * F_{r+2} * \cdots * F_n \in \mathcal{L}_{2^{n-r} \times 2^n},
 \end{aligned}$$

and

$$H = H_1 * H_2 * \cdots * H_p \in \mathcal{L}_{2^p \times 2^n}.$$

Summarizing, we obtain the following algebraic form of system (6):

$$\begin{cases}
 x(t + 1) = Qu(t)x(t), \\
 y(t) = Hx(t),
 \end{cases}
 \tag{10}$$

where $Q = L_1(I_{2^{n+r}} \otimes L_2)W_{[2^n, 2^{n+r}]}M_{r, 2^n}W_{[2^r, 2^n]} \in \mathcal{L}_{2^n \times 2^{n+r}}$.

Now, we consider system (10) with state and input constraints. For any $t \in \mathbb{N}$, we assume that $x(t) \in S_x \subseteq \Delta_{2^n}$ and $u(t) \in S_u \subseteq \Delta_{2^r}$. Let $|S_x| = n_1 \leq 2^n$ and $|S_u| = r_1 \leq 2^r$, where $|S_x|$ denotes the cardinality of the set S_x . Then, S_x and S_u can be expressed as

$$S_x = \{\delta_{2^n}^{i_k} : k = 1, \dots, n_1; 1 \leq i_1 < \cdots < i_{n_1} \leq 2^n\},
 \tag{11}$$

$$S_u = \{\delta_{2^r}^{j_k} : k = 1, \dots, r_1; 1 \leq j_1 < \cdots < j_{r_1} \leq 2^r\}.
 \tag{12}$$

Denote the trajectory of system (10) with a pinning control sequence $\{(u_1(t), u_2(t), \dots, u_r(t)) : t \in \mathbb{N}\} \subseteq S_u$ and an initial state $x_0 \in S_x$ by $x(t; x_0, (u_1(t), u_2(t), \dots, u_r(t)))$.

In the following, we convert system (10) with state and input constraints into an equivalent constrained algebraic form.

Define the following set of matrices:

$$J_i^{(p,q)} := \underbrace{[0_{q \times q} \quad \cdots \quad 0_{q \times q} \quad \underbrace{I_q}_{i\text{th}} \quad 0_{q \times q} \quad \cdots \quad 0_{q \times q}]}_p,
 \tag{13}$$

where $J_i^{(p,q)} \in \mathbb{R}^{q \times pq}$, $i = 1, 2, \dots, p$, $0_{q \times q}$ denotes the $q \times q$ zero matrix, and $I_q \in \mathcal{L}_{q \times q}$ is the $q \times q$ identity matrix.

Proposition 4 ([35]) 1. Given a matrix $A \in \mathbb{R}^{pq \times r}$, split A as

$$A = \begin{bmatrix} A_1 \\ \vdots \\ A_p \end{bmatrix}, \tag{14}$$

where $A_i \in \mathbb{R}^{q \times r}$. Then,

$$J_i^{(p,q)} A = A_i. \tag{15}$$

2. Given a matrix $B \in \mathbb{R}^{r \times pq}$, split B as

$$B = [B_1 \quad \cdots \quad B_p], \tag{16}$$

where $B_i \in \mathbb{R}^{r \times q}$. Then,

$$B(J_i^{(p,q)})^T = B_i. \tag{17}$$

Based on Proposition 4, set

$$\Psi_x = \begin{bmatrix} J_{i_1}^{(2^n,1)} \\ \vdots \\ J_{i_{n_1}}^{(2^n,1)} \end{bmatrix}, \tag{18}$$

$$\Psi_u = \begin{bmatrix} J_{j_1}^{(2^r,1)} \\ \vdots \\ J_{j_{r_1}}^{(2^r,1)} \end{bmatrix}. \tag{19}$$

Denote $\delta_{n_1}^0 = 0_{n_1 \times 1}$ and $\delta_{r_1}^0 = 0_{r_1 \times 1}$. Then, we convert the state $x(t) \in \Delta_{2^n}$ and input $u(t) \in \Delta_{2^r}$ of the constrained system into the following form:

$$\widehat{x}(t) = \Psi_x x(t) \in \widehat{\mathcal{S}}_x, \tag{20}$$

$$\widehat{u}(t) = \Psi_u u(t) \in \widehat{\mathcal{S}}_u, \tag{21}$$

where $\widehat{\mathcal{S}}_x = \{\delta_{n_1}^1, \delta_{n_1}^2, \dots, \delta_{n_1}^{n_1}\} \cup \{\delta_{n_1}^0\}$ and $\widehat{\mathcal{S}}_u = \{\delta_{r_1}^1, \delta_{r_1}^2, \dots, \delta_{r_1}^{r_1}\} \cup \{\delta_{r_1}^0\}$.

For system (10), let $Q = [Q_1, \dots, Q_{2^r}]$, $Q_l \in \mathcal{L}_{2^n \times 2^n}$, $l \in \{1, \dots, 2^r\}$. Set

$$\widehat{Q} = [\widehat{Q}_1 \quad \cdots \quad \widehat{Q}_{2^r}] \times [(J_{j_1}^{(2^r, m_1)})^T \quad \cdots \quad (J_{j_{r_1}}^{(2^r, m_1)})^T] \in \mathcal{B}_{n_1 \times n_1 r_1}, \tag{22}$$

$$\widehat{H} = H \Psi_x^T \in \mathcal{L}_{2^p \times n_1}, \tag{23}$$

where

$$\widehat{Q}_l = \Psi_x Q_l \Psi_x^T \in \mathcal{B}_{n_1 \times n_1}, \quad l \in \{1, \dots, 2^r\}. \tag{24}$$

Remark 1 Since Q is a logical matrix, we can easily conclude that each column of \widehat{Q} has at most one element '1'.

Based on the above transformation, we convert system (10) into the following form:

$$\begin{cases} \widehat{x}(t+1) = \widehat{Q}\widehat{u}(t)\widehat{x}(t), \\ \widehat{y}(t) = \widehat{H}\widehat{x}(t). \end{cases} \tag{25}$$

Proposition 5 *The state trajectories of system (10) with S_x and S_u are equivalent to that of system (25) with \widehat{S}_x and \widehat{S}_u .*

Proof On one hand, $\forall t \in \mathbb{N}, \forall u(t) = \delta_{2^r}^{j_s} \in S_u$, and $\forall x(t) = \delta_{2^n}^{i_\alpha} \in S_x$, if $x(t+1) = Qu(t)x(t) \in S_x$, say, $x(t+1) = \delta_{2^n}^{i_\beta}$, $\beta \in \{1, \dots, n_1\}$, a simple calculation shows that

$$\widehat{x}(t+1) = \widehat{Q}\widehat{u}(t)\widehat{x}(t) = \delta_{n_1}^\beta \in \widehat{S}_x \setminus \{\delta_{n_1}^0\},$$

where $\widehat{x}(t) = \delta_{n_1}^\alpha \in \widehat{S}_x$ and $\widehat{u}(t) = \delta_{r_1}^s \in \widehat{S}_u$. If $x(t+1) = Qu(t)x(t) = \delta_{2^n}^{i_\alpha} \notin S_x$, then $\widehat{x}(t+1) = \widehat{Q}\widehat{u}(t)\widehat{x}(t) = \delta_{n_1}^0$. Hence, in both cases, $\widehat{x}(t+1) = \Psi_x x(t+1)$.

On the other hand, $\forall t \in \mathbb{N}, \forall \widehat{u}(t) = \delta_{r_1}^s \in \widehat{S}_u \setminus \{\delta_{r_1}^0\}$, and $\forall \widehat{x}(t) = \delta_{n_1}^\alpha \in \widehat{S}_x \setminus \{\delta_{n_1}^0\}$, if $\widehat{x}(t+1) = \widehat{Q}\widehat{u}(t)\widehat{x}(t) \in \widehat{S}_x \setminus \{\delta_{n_1}^0\}$, say, $\widehat{x}(t+1) = \delta_{n_1}^\beta$, $\beta \in \{1, \dots, n_1\}$, we can obtain that

$$x(t+1) = Qu(t)x(t) = \delta_{2^n}^{i_\beta} \in S_x.$$

If $\widehat{x}(t+1) = \widehat{Q}\widehat{u}(t)\widehat{x}(t) = \delta_{n_1}^0$, then

$$x(t+1) = Qu(t)x(t) \notin S_x.$$

Hence, we have $\widehat{x}(t+1) = \Psi_x x(t+1)$.

Therefore, the state trajectories of system (10) with S_x and S_u are equivalent to that of system (25) with \widehat{S}_x and \widehat{S}_u . □

Remark 2 We call (25) the constrained algebraic form of the original system. Based on Proposition 5, we can convert the feedback stabilization of the original system to that of system (25).

3.2 Reachability analysis

In this subsection, we study the reachability of system (25), which is crucial to pinning control design for the feedback stabilization.

We give the definition of the reachability for system (25) as follows.

Definition 2 For system (25), given two states $\widehat{x}_0, \widehat{x}_d \in \widehat{S}_x \setminus \{\delta_{n_1}^0\}$ and a integer $k > 0$, \widehat{x}_d is said to be reachable from \widehat{x}_0 at time k if there is a pinning control sequence $(\widehat{u}_1(t), \widehat{u}_2(t), \dots, \widehat{u}_r(t))$ with $\widehat{u}(t) = \times_{i=1}^r \widehat{u}_i(t) \in \widehat{S}_u \setminus \{\delta_{r_1}^0\}$, $t \in \{1, 2, \dots, k-1\}$, such that $\widehat{x}(k; \widehat{x}_0, (\widehat{u}_1(t), \widehat{u}_2(t), \dots, \widehat{u}_r(t))) = \widehat{x}_d$.

For system (25), consider two given states $\widehat{x}_0 = \delta_{n_1}^\alpha \in \widehat{S}_x \setminus \{\delta_{n_1}^0\}$, $\widehat{x}_d = \delta_{n_1}^\beta \in \widehat{S}_x \setminus \{\delta_{n_1}^0\}$ and a given integer $k > 0$. Let $P(k; \widehat{x}_0, \widehat{x}_d)$ denote the number of different paths such that \widehat{x}_d is reachable from \widehat{x}_0 at time k .

Lemma 2 Consider system (25) with two given states $\widehat{x}_0 = \delta_{n_1}^\alpha \in \widehat{S}_x \setminus \{\delta_{n_1}^0\}$, $\widehat{x}_d = \delta_{n_1}^\beta \in \widehat{S}_x \setminus \{\delta_{n_1}^0\}$ and a given integer $k > 0$. Then,

$$P(k; \widehat{x}_0, \widehat{x}_d) = (\widehat{x}_d^T)(\overline{Q})^k(\widehat{x}_0), \tag{26}$$

where $\overline{Q} = \sum_{i=1}^{r_1} \widehat{Q}_i$, and \widehat{Q}_i is defined in (24).

Proof We prove this lemma by induction.

Firstly, letting $k = 1$, assume that $\widehat{u}^1 = \delta_{r_1}^{\rho_1}, \widehat{u}^2 = \delta_{r_1}^{\rho_2}, \dots, \widehat{u}^s = \delta_{r_1}^{\rho_s} \in \widehat{S}_u \setminus \{\delta_{r_1}^0\}$ are different control sequences such that \widehat{x}_d is reachable from \widehat{x}_0 at one step. Let $\widehat{u}^{s+1} = \delta_{r_1}^{\rho_{s+1}}, \widehat{u}^{s+2} = \delta_{r_1}^{\rho_{s+2}}, \dots, \widehat{u}^{r_1} = \delta_{r_1}^{\rho_{r_1}} \in \widehat{S}_u \setminus \{\delta_{r_1}^0\}$ be different control sequences such that \widehat{x}_0 cannot reach \widehat{x}_d in one step. Hence, it is easy to see that $(\widehat{Q}_{\rho_l})_{\beta,\alpha} = 1, \forall l \in \{1, \dots, s\}$, and $(\widehat{Q}_{\rho_l})_{\beta,\alpha} = 0, \forall l \in \{s+1, \dots, r_1\}$, which implies that $(\sum_{i=1}^{r_1} \widehat{Q}_i)_{\beta,\alpha} = s$. Since $(\widehat{x}_d^T)(\overline{Q})(\widehat{x}_0) = (\sum_{i=1}^{r_1} \widehat{Q}_i)_{\beta,\alpha}$, we have

$$P(1; \widehat{x}_0, \widehat{x}_d) = s = (\widehat{x}_d^T)(\overline{Q})(\widehat{x}_0).$$

Thus, (26) holds for $k = 1$.

Suppose that (26) holds for an integer $k \geq 1$. Then, we consider the case of $k + 1$. It is easy to see that

$$\begin{aligned} \widehat{x}_d^T(\overline{Q})^{k+1}\widehat{x}_0 &= (\delta_{n_1}^\beta)^T(\overline{Q})^{k+1}\delta_{n_1}^\alpha = (\overline{Q}^k\overline{Q})_{(\beta,\alpha)} \\ &= \sum_{p=1}^{n_1} (\delta_{n_1}^\beta)^T(\overline{Q}^k)(\delta_{n_1}^p)(\delta_{n_1}^p)^T(\overline{Q})\delta_{n_1}^\alpha \\ &= \sum_{p=1}^{n_1} P(k; \delta_{n_1}^p, \widehat{x}_d)P(1; \widehat{x}_0, \delta_{n_1}^p) \\ &= P(k+1; \widehat{x}_0, \widehat{x}_d), \end{aligned} \tag{27}$$

which shows that (26) holds for $k + 1$.

By induction, (26) holds for any integer $k > 0$. This completes the proof. □

Based on Lemma 2, we give the following result on the reachability of system (25).

Theorem 1 For system (25), $\widehat{x}_d = \delta_{n_1}^\beta \in \widehat{S}_x \setminus \{\delta_{n_1}^0\}$ is reachable from $\widehat{x}_0 = \delta_{n_1}^\alpha \in \widehat{S}_x \setminus \{\delta_{n_1}^0\}$ at time k if and only if

$$(\overline{Q}^k)_{\beta,\alpha} > 0.$$

Proof By Lemma 2 we get $P(k; \widehat{x}_0, \widehat{x}_d) = (\delta_{n_1}^\beta)^T(\overline{Q})^k(\delta_{n_1}^\alpha) = (\overline{Q}^k)_{\beta,\alpha}$. Therefore, $P(k; \widehat{x}_0, \widehat{x}_d)$ denotes the number of different paths from \widehat{x}_0 to \widehat{x}_d at time k . If $(\overline{Q}^k)_{\beta,\alpha} > 0$, then there are $(\overline{Q}^k)_{\beta,\alpha}$ different paths from \widehat{x}_0 to \widehat{x}_d . Thus, \widehat{x}_d can be reached from \widehat{x}_0 at time k . Conversely, if \widehat{x}_d is reachable from \widehat{x}_0 at time k , then we can find at least one pinning control sequence $(\widehat{u}_1(t), \widehat{u}_2(t), \dots, \widehat{u}_r(t))$ with $\widehat{u}(t) = \times_{i=1}^r \widehat{u}_i(t) \in \widehat{S}_u \setminus \{\delta_{r_1}^0\}, t \in \{1, 2, \dots, k-1\}$, such that $\widehat{x}(k; \widehat{x}_0, (\widehat{u}_1(t), \widehat{u}_2(t), \dots, \widehat{u}_r(t))) = \widehat{x}_d$, which implies that $(\overline{Q}^k)_{\beta,\alpha} = P(k; \widehat{x}_0, \widehat{x}_d) > 0$. □

Remark 3 From the proof of Theorem 1 we easily see that

$$P(k; \widehat{x}_0, \widehat{x}_d) = (\delta_{n_1}^\beta)^T (\overline{Q})^k (\delta_{n_1}^\alpha) = (\overline{Q}^k)_{\beta, \alpha}.$$

All of them denote the number of different paths such that \widehat{x}_d is reachable from \widehat{x}_0 at time k .

3.3 Feedback stabilization pinning control design

In this part, we study the pinning control design for the feedback stabilization of constrained BCNs. By Proposition 5 we consider the feedback stabilization of system (25) based on the reachability analysis.

Firstly, we give the definition of stabilization for system (10) with S_x and S_u .

Definition 3 System (10) with S_x and S_u is said to be stabilizable to a given equilibrium $x_e \in S_x$ if there exists a pinning control sequence $\{(u_1(t), u_2(t), \dots, u_r(t)) : t \in \mathbb{N}\} \subset S_u$ under which the trajectory initialized at any $x_0 \in S_x$ converges to x_e and $x(t; x_0, (u_1(t), u_2(t), \dots, u_r(t))) \in S_x, \forall t \in \mathbb{N}$.

In this paper, we study the following two kinds of feedback pinning controls:

1. State feedback pinning control:

$$u_i(t) = K_i x(t), \tag{28}$$

where $K_i \in \mathcal{L}_{2 \times 2^n}, i = 1, \dots, r$.

2. Output feedback pinning control:

$$u_i(t) = G_i y(t), \tag{29}$$

where $G_i \in \mathcal{L}_{2 \times 2^p}, i = 1, \dots, r$.

In the following, we consider the pinning control design for the state feedback stabilization of system (10) with S_x and S_u based on the constrained algebraic form.

For system (25), let $\widehat{x}_e = \Psi_x x_e = \delta_{n_1}^\alpha \in \widehat{S}_x \setminus \{\delta_{n_1}^0\}$. For any integer $k > 0$, define

$$\begin{aligned} \Lambda_k(\widehat{x}_e) &= \{ \delta_{n_1}^\beta \in \widehat{S}_x \setminus \{ \delta_{n_1}^0 \} : \text{there exists a control sequence} \\ &\quad \widehat{u}(0), \widehat{u}(1), \dots, \widehat{u}(k-1) \in \widehat{S}_u \setminus \{ \delta_{n_1}^0 \} \text{ such that} \\ &\quad \widehat{x}(k; \delta_{n_1}^\beta, \widehat{u}(0), \dots, \widehat{u}(k-1)) = \delta_{n_1}^\alpha \text{ and} \\ &\quad \widehat{x}(l; \delta_{n_1}^\beta, \widehat{u}(0), \dots, \widehat{u}(l-1)) \in \widehat{S}_x \setminus \{ \delta_{n_1}^0 \}, \forall l \in \{1, \dots, k\} \}. \end{aligned} \tag{30}$$

Proposition 6 System (25) is stabilized to $\widehat{x}_e = \delta_{n_1}^\alpha$ by a state feedback control if and only if there exists a positive integer $\sigma \leq n_1$ such that

1. $(\overline{Q})_{\alpha, \alpha} > 0$,
2. $\text{Row}_\alpha(\overline{Q}^\sigma) > 0$.

Proof (Sufficiency) We can see from Condition 1 and Theorem 1 that $\widehat{x}_e \in \Lambda_1(\widehat{x}_e)$, which implies that $\Lambda_k(\widehat{x}_e) \neq \emptyset, \forall k = 1, \dots, \sigma$.

Denote

$$\Lambda_k^\circ(\widehat{x}_e) = \Lambda_k(\widehat{x}_e) \setminus \Lambda_{k-1}(\widehat{x}_e), \tag{31}$$

where $\Lambda_0(\widehat{x}_e) := \emptyset$. Then, we obtain $\Lambda_{k_1}^\circ(\widehat{x}_e) \cap \Lambda_{k_2}^\circ(\widehat{x}_e) = \emptyset, \forall k_1, k_2 \in \{1, \dots, \sigma\}, k_1 \neq k_2$. Moreover, by Condition 2 and Theorem 1 we have $\bigcup_{k=1}^\sigma \Lambda_k^\circ(\widehat{x}_e) = \widehat{S}_x \setminus \{\delta_{n_1}^0\}$. Thus, for any integer i satisfying $1 \leq i \leq n_1$, we can find the unique integer $1 \leq k_i \leq \sigma$ such that $\delta_{n_1}^i \in \Lambda_{k_i}^\circ(\widehat{x}_e)$.

For system (25), set $\widehat{Q} = \delta_{n_1}[\mu_1, \mu_2, \dots, \mu_{n_1 r_1}]$. When $k_i = 1$, we can find an integer $1 \leq \omega_i \leq r_1$ such that $\widehat{Q} \times \delta_{r_1}^{\omega_i} \times \delta_{n_1}^i = \delta_{n_1}^{\mu(\omega_i-1)n_1+i} = \widehat{x}_e$. When $2 \leq k_i \leq \sigma$, we can find an integer $1 \leq \omega_i \leq r_1$ such that $\widehat{Q} \times \delta_{r_1}^{\omega_i} \times \delta_{n_1}^i = \delta_{n_1}^{\mu(\omega_i-1)n_1+i} \in \Lambda_{k_i-1}(\widehat{x}_e)$.

Let

$$\widehat{K} = \delta_{r_1}[\omega_1, \omega_2, \dots, \omega_{n_1}] \in \mathcal{L}_{r_1 \times n_1}. \tag{32}$$

Then, for any initial state $\widehat{x}_0 = \delta_{n_1}^i \in \widehat{S}_x \setminus \{\delta_{n_1}^0\}$, if $k_i = 1$, then we obtain

$$\widehat{x}(1; \widehat{x}_0, \widehat{u}) = \widehat{Q}\widehat{u}\widehat{x}_0 = \widehat{Q}\widehat{K}\widehat{x}_0\widehat{x}_0 = \delta_{n_1}^{\mu(\omega_i-1)n_1+i} = \widehat{x}_e;$$

if $2 \leq k_i \leq \sigma$, we have

$$\widehat{x}(1; \widehat{x}_0, \widehat{u}) = \widehat{Q}\widehat{u}\widehat{x}_0 = \widehat{Q}\widehat{K}\widehat{x}_0\widehat{x}_0 = \delta_{n_1}^{\mu(\omega_i-1)n_1+i} \in \Lambda_{k_i-1}(\widehat{x}_e).$$

Hence, $\widehat{x}(k_i; \widehat{x}_0, \widehat{u}) = \widehat{x}_e, \forall 1 \leq i \leq n_1$, and $\widehat{x}(t; \widehat{x}_0, \widehat{u}) \in \widehat{S}_x \setminus \{\delta_{n_1}^0\}, \forall 0 \leq t \leq k_i - 1$. Since $\widehat{x}_e \in \Lambda_1(\widehat{x}_e)$, we obtain

$$\widehat{x}(t; \widehat{x}_0, \widehat{u}) = \widehat{x}_e, \quad \forall t \geq \sigma, \forall \widehat{x}_0 \in \widehat{S}_x \setminus \{\delta_{n_1}^0\},$$

which implies that system (25) is stabilized to $\widehat{x}_e = \delta_{n_1}^\alpha$ by the state feedback control $\widehat{u}(t) = \widehat{K}\widehat{x}(t)$.

(Necessity) The proof of this part is based on a straightforward calculation, and thus we omit it here. □

Based on Propositions 5 and 6, we have the following result.

Theorem 2 *System (6) with S_x and S_u is stabilized to x_e by a state feedback control if and only if there exists a positive integer $\sigma \leq n_1$ such that $(\overline{Q})_{\alpha, \alpha} > 0$ and $\text{Row}_\alpha(\overline{Q}^\sigma) > 0$.*

From the proof of Proposition 6 we get the following procedure for the pinning control design of state feedback stabilization of constrained BCNs.

Remark 4 The procedure contains the following steps:

1. Calculate $\Lambda_k(\widehat{x}_e)$ and $\Lambda_k^\circ(\widehat{x}_e), k = 1, \dots, \sigma$.
2. For every integer $1 \leq i \leq n_1$, find the unique integer $1 \leq k_i \leq \sigma$ satisfying $\delta_{n_1}^i \in \Lambda_{k_i}^\circ(\widehat{x}_e)$.
3. Find an integer $1 \leq \omega_i \leq r_1$ such that if $k_i = 1$, then $\delta_{n_1}^{\mu(\omega_i-1)n_1+i} = \widehat{x}_e$; if $k_i \geq 2$, then $\delta_{n_1}^{\mu(\omega_i-1)n_1+i} \in \Lambda_{k_i-1}(\widehat{x}_e)$.

4. The state feedback pinning control can be designed as $u_i(t) = K_i x(t)$, $i = 1, \dots, r$, with $K_1 * K_2 * \dots * K_r = K$, where $K = \delta_{2^r} [p_1, \dots, p_{2^r}]$, and

$$\begin{cases} p_t = j_{\omega_\rho} & \text{if } t = i_\rho, \rho \in \{1, \dots, n_1\}, \\ p_t \in \{j_1, \dots, j_{r_1}\} & \text{otherwise.} \end{cases} \tag{33}$$

Finally, we discuss the pinning control design for the output feedback stabilization of constrained BCNs. To this end, we recall the definition of a nilpotent matrix.

Definition 4 A nilpotent matrix N is a square matrix such that $N^k = 0$ for some positive integer k . The smallest such k is called the degree of N .

For system (25) with an output feedback control $\hat{u}(t) = \widehat{G}\hat{y}(t)$, $\widehat{G} \in \mathcal{L}_{r_1 \times 2^p}$, we have

$$\hat{x}(t + 1) = \widehat{Q}\hat{u}(t)\hat{x}(t) = \widehat{Q}\widehat{G}\hat{y}(t)\hat{x}(t) = \widehat{Q}\widehat{G}\widehat{H}M_{r,n_1}\hat{x}(t), \tag{34}$$

where $M_{r,n_1} = \text{diag}\{\delta_{n_1}^1, \delta_{n_1}^2, \dots, \delta_{n_1}^{n_1}\}$. Then, we have the following result on the output feedback stabilization of system (25).

Theorem 3 System (25) is stabilizable to $\hat{x}_e = \delta_{n_1}^\alpha$ by an output feedback control if and only if there exist a logical matrix $\widehat{G} \in \mathcal{L}_{r_1 \times 2^p}$ and an integer $1 \leq \tau \leq n_1$ such that

$$\widehat{Q}\widehat{G}\widehat{H}M_{r,n_1} = \begin{bmatrix} A_1 & \zeta_3 & A_2 \\ \zeta_1 & 1 & \zeta_2 \\ A_3 & \zeta_4 & A_4 \end{bmatrix} \tag{35}$$

and

$$\begin{bmatrix} A_1 & A_2 \\ A_3 & A_4 \end{bmatrix} \tag{36}$$

is a nilpotent matrix of degree τ , where $A_1 \in \mathcal{B}_{(\alpha-1) \times (\alpha-1)}$, $A_2 \in \mathcal{B}_{(\alpha-1) \times (n_1-\alpha)}$, $A_3 \in \mathcal{B}_{(n_1-\alpha) \times (\alpha-1)}$, $A_4 \in \mathcal{B}_{(n_1-\alpha) \times (n_1-\alpha)}$, ζ_1 and ζ_2 are some proper Boolean row vectors, and ζ_3 and ζ_4 are zero column vectors.

Proof (Sufficiency) Since

$$\begin{bmatrix} A_1 & A_2 \\ A_3 & A_4 \end{bmatrix} \tag{37}$$

is a nilpotent matrix of degree τ , we see that

$$\begin{bmatrix} A_1 & A_2 \\ A_3 & A_4 \end{bmatrix}^t = 0 \tag{38}$$

for any integer $t \geq \tau$, which, together with a simple calculation, shows that

$$(\widehat{Q}\widehat{G}\widehat{H}M_{r,n_1})^t = \begin{bmatrix} 0 & 0 & \cdots & 0 \\ 1 & 1 & \cdots & 1 \\ 0 & 0 & \cdots & 0 \end{bmatrix} = \delta_{n_1}[\alpha, \alpha, \dots, \alpha] \tag{39}$$

for any integer $t \geq \tau$.

Hence, we obtain from (34) that

$$\widehat{x}(t) = (\widehat{Q}\widehat{G}\widehat{H}M_{r,n_1})^t \widehat{x}(0) = \widehat{x}_e \tag{40}$$

for any $\widehat{x}(0) \in \widehat{S}_x \setminus \{\delta_{n_1}^0\}$ and any integer $t \geq \tau$.

Therefore, system (25) is stabilizable to $\widehat{x}_e = \delta_{n_1}^\alpha$ by the output feedback control $\widehat{u}(t) = \widehat{G}\widehat{y}(t)$.

(Necessity) Suppose that system (25) is stabilizable to $\widehat{x}_e = \delta_{n_1}^\alpha$ by an output feedback control, say, $\widehat{u}(t) = \widehat{G}\widehat{y}(t)$. Then, we can find the smallest integer $1 \leq \tau \leq n_1$ such that (40) holds for any $\widehat{x}(0) \in \widehat{S}_x \setminus \{\delta_{n_1}^0\}$ and any integer $t \geq \tau$. Hence, $(\widehat{Q}\widehat{G}\widehat{H}M_{r,n_1})^\tau = \delta_{n_1}[\alpha, \alpha, \dots, \alpha]$.

Split $\widehat{Q}\widehat{G}\widehat{H}M_{r,n_1} \in \mathcal{L}_{n_1 \times n_1}$ into the following blocks:

$$\widehat{Q}\widehat{G}\widehat{H}M_{r,n_1} = \begin{bmatrix} A_1 & \zeta_3 & A_2 \\ \zeta_1 & \lambda & \zeta_2 \\ A_3 & \zeta_4 & A_4 \end{bmatrix},$$

where $\lambda \in \{0, 1\}$, $A_1 \in \mathcal{B}_{(\alpha-1) \times (\alpha-1)}$, $A_2 \in \mathcal{B}_{(\alpha-1) \times (n_1-\alpha)}$, $A_3 \in \mathcal{B}_{(n_1-\alpha) \times (\alpha-1)}$, $A_4 \in \mathcal{B}_{(n_1-\alpha) \times (n_1-\alpha)}$, ζ_1 and ζ_2 are some proper Boolean row vectors, and ζ_3 and ζ_4 are some proper Boolean column vectors.

It is easy to see from $\widehat{Q}\widehat{G}\widehat{H}M_{r,n_1}\widehat{x}_e = \widehat{x}_e$ that $\lambda = 1$, and ζ_3 and ζ_4 are zero column vectors.

In the following, we prove that

$$\widehat{A} := \begin{bmatrix} A_1 & A_2 \\ A_3 & A_4 \end{bmatrix}$$

is a nilpotent matrix of degree τ .

We can easily see from $(\widehat{Q}\widehat{G}\widehat{H}M_{r,n_1})^\tau = \delta_{n_1}[\alpha, \alpha, \dots, \alpha]$ that $\widehat{A}^\tau = 0$, which shows that \widehat{A} is a nilpotent matrix. If its degree is less than τ , then there exists a positive integer $\tau' < \tau$ such that $\widehat{A}^{\tau'} = 0$, and thus

$$(\widehat{Q}\widehat{G}\widehat{H}M_{r,n_1})^{\tau'} = \delta_{n_1}[\alpha, \alpha, \dots, \alpha],$$

which is a contradiction to the minimality of τ . This completes the proof. □

Remark 5 Based on Theorem 3 and Algorithm 1 presented in [32], we can design an output feedback gain matrix, say, $\widehat{G} = \delta_{r_1}[w_1, w_2, \dots, w_{2^p}]$, under which system (25) is stabilizable to $\widehat{x}_e = \delta_{n_1}^\alpha$. Then, the output feedback pinning control can be designed as $u_i(t) = G_i y(t)$, $i = 1, \dots, r$, with $G_1 * G_2 * \dots * G_r = G$, where $G = \delta_{2^r}[j_{w_1}, \dots, j_{w_{2^p}}]$.

Remark 6 It should be pointed out that the pinning control design for the state feedback stabilization of Boolean networks was studied in [34], and a novel design procedure was established. Compared with [34], our main results have the following advantages: (i) we established some necessary and sufficient conditions for the pinning control design of both state feedback and output feedback stabilization problems, whereas [34] only considered the state feedback stabilization problem; (ii) our results are applicable to the pinning control design for the state feedback stabilization of constrained BCNs.

4 Illustrative examples

In the section, we give two illustrative examples to show how to use the obtained results to design pinning control for feedback stabilization of constrained BCNs.

Example 1 Consider the following Boolean model to simulate the λ phage, which is a virus growing on a bacterium. The virus can only follow one of two different pathways: lysogeny or lysis, after injecting chromosome into the bacterium cell. The molecular mechanism responsible for the lysogeny/lysis decision is known as λ switch [36]. For example, two genes, *cI* and *cro*, directly affect the decision. When *cI* is active (inactive) and *cro* is inactive (active), the phage is in the lysogenic (lytic) state, and whether lysogenic state will be established or not depends on five phage genes, *cI*, *cro*, *cII*, *cIII*, *N*, and the environmental state. More details can be found in [37]. Its dynamic can be described in the form

$$\begin{cases} cII(t + 1) = (\neg cI(t)) \wedge (N(t) \vee cIII(t)) \wedge u(t), \\ cIII(t + 1) = (\neg cI(t)) \wedge N(t) \wedge u(t), \\ N(t + 1) = (\neg cI(t)) \wedge (\neg cro(t)), \\ cI(t + 1) = (\neg cro(t)) \wedge (cI(t) \vee cII(t)), \\ cro(t + 1) = (\neg cI(t)) \wedge (\neg cII(t)), \end{cases} \tag{41}$$

where $u(t)$ is a binary input that represents whether environmental condition is favorable or not.

Letting $x(t) = cII(t)cIII(t)N(t)cI(t)cro(t)$, we obtain the following algebraic form:

$$x(t + 1) = Qu(t)x(t), \tag{42}$$

where

$$Q = \delta_{32} [\begin{matrix} 32 & 30 & 8 & 2 & 32 & 30 & 16 & 10 & 32 & 30 & 8 & 2 & 32 & 30 & 16 & 10 \\ 32 & 30 & 7 & 3 & 32 & 30 & 31 & 27 & 32 & 30 & 7 & 3 & 32 & 30 & 31 & 27 \\ 32 & 30 & 32 & 26 & 32 & 30 & 32 & 26 & 32 & 30 & 32 & 26 & 32 & 30 & 32 & 26 \\ 32 & 30 & 31 & 27 & 32 & 30 & 31 & 27 & 32 & 30 & 31 & 27 & 32 & 30 & 31 & 27 \end{matrix}].$$

In this example, several environmental conditions including concentration of nutrition, growth rate, temperature, and multiplicity of infection can influence the *cII* and *cIII* genes. If the environmental conditions are favorable, then the *cII* and *cIII* genes are highly active, and the *cII* gene product turns the *cI* gene on. The *cI* gene inhibits all other genes including *cro*, and the lysogenic state is established. If the environmental conditions are not

favorable, then the genes cII and cIII are not activated, the cro gene remains active, and its product represses the cI gene. Thus, the lytic state is established.

Due to the limitation of the environmental conditions, we constrain the state in $S_x = \{\delta_{32}^{19}, \delta_{32}^{23}, \delta_{32}^{27}, \delta_{32}^{31}\}$.

By Theorem 2 and Remark 4 we can design 2^{30} state feedback pinning controls that stabilize system (41) with S_x to the equilibrium δ_{32}^{31} (the lytic state), and one of them is

$$u(t) = \delta_2[1\ 2\ 1\ 2\ 1\ 2\ 1\ 2\ 1\ 2\ 1\ 2\ 1\ 2\ 1\ 2\ 1\ 2\ 2\ 1\ 1\ 2\ 2\ 1\ 1\ 2\ 2\ 1\ 1\ 2\ 2\ 1]x(t).$$

Example 2 Consider the following BCN:

$$\begin{cases} x_1(t + 1) = x_2(t) \vee u_1(t), \\ x_2(t + 1) = (x_3(t) \rightarrow x_1(t)) \vee u_2(t), \\ x_3(t + 1) = \neg x_3(t); \\ y_1(t) = x_2(t) \vee x_3(t), \\ y_2(t) = (x_1(t) \wedge x_2(t)) \vee (\neg x_1(t) \wedge (x_2(t) \leftrightarrow x_3(t))). \end{cases} \tag{43}$$

Setting $x(t) = \times_{i=1}^3 x_i(t)$, $u(t) = \times_{i=1}^2 u_i(t)$, and $y(t) = \times_{i=1}^2 y_i(t)$, we obtain the following algebraic form:

$$\begin{cases} x(t + 1) = Qu(t)x(t), \\ y(t) = Hx(t), \end{cases} \tag{44}$$

where

$$Q = \delta_8[2\ 2\ 2\ 2\ 1\ 1\ 1\ 1\ 2\ 2\ 6\ 6\ 1\ 1\ 5\ 5\ 2\ 4\ 2\ 4\ 1\ 1\ 1\ 1\ 2\ 4\ 6\ 8\ 1\ 1\ 5\ 5]$$

and

$$H = \delta_4[1\ 1\ 2\ 4\ 1\ 2\ 2\ 3].$$

We assume that $S_x = \{\delta_8^1, \delta_8^2, \delta_8^3, \delta_8^4, \delta_8^5, \delta_8^8\}$, and $S_u = \{\delta_4^1, \delta_4^2, \delta_4^3, \delta_4^4\}$. We aim to design an output feedback pinning control such that system (43) with S_x and S_u is stabilized to $x_e = \delta_8^2$.

Based on Theorem 3 and Algorithm 1 presented in [32], we can obtain eight output feedback pinning control gain matrices; one of them is $K_1 = \delta_2[1\ 1\ 2\ 1]$, $K_2 = \delta_2[1\ 1\ 1\ 1]$.

5 Conclusion

In this paper, we have studied the pinning control design for feedback stabilization of constrained BCNs. We have obtained the constrained algebraic form for constrained BCNs with pinning control via the semitensor product of matrices. We have given a necessary and sufficient condition for the reachability of constrained BCNs. We have proposed a procedure for the pinning control design of state feedback stabilization of constrained BCNs. Moreover, we have presented a necessary and sufficient condition for the output feedback stabilization of constrained BCNs with pinning control. The study of two illustrative examples has shown that the new results obtained in this paper are very effective.

Competing interests

The authors declare that they have no competing interests.

Authors' contributions

QY and HL carried out the main part of this article, and YL corrected the manuscript and brought forward many suggestions on this article. All authors have read and approved the final manuscript.

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